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Poirot Modeling Design Doc

gTrade team

General Information

1. What does the model predict?

The model generates the bid reduction to be applied to DBM fixed CPM bids to maximize advertiser surplus.

2. How and where is the prediction used?

The bid reduction is applied on the fixed CPM bids before the internal DBM auction is run.

3. Contact information (team/TL email, wiki page).

gTrade. go/gtrade, gtrade-core@google.com

Source of Training Data

1. How is the training data obtained?

[REDACTED]

2. Does it correctly handle conversion data (see Conversion Handling)?

No conversion data used.

3. Does it correctly handle DBM data (see DBM Data Separation)?

Yes. This pipeline only operates on DBM data.

4. How are bad data handled (e.g., when certain days of data are deemed unusable)?

We can easily exclude bad days through the borg files.

5. Are the training data stored?

- a. If so, does it satisfy wipeout compliance?

No user information is stored.

6. How frequently does the training data get refreshed?

Every day.

7. What happens when training data is stale?

We get alerts when any job files or the model file is stale.

Training Algorithm

1. How is the model trained?

Commented [1]: so this only works for fixed cpm? how about opt advertisers? and if not, why not?

Commented [2]: For opt, Ali wants to combine it with HDML. We have to think about it. Fixed is 90% of DBM anyway, so we are ok with this in v1

Commented [3]: does this rely on running long standing holdback experiments?

Commented [4]: yes. we have to explore

Commented [5]: Eujin's suggestion: do this in the code directly rather than using an experiment framework

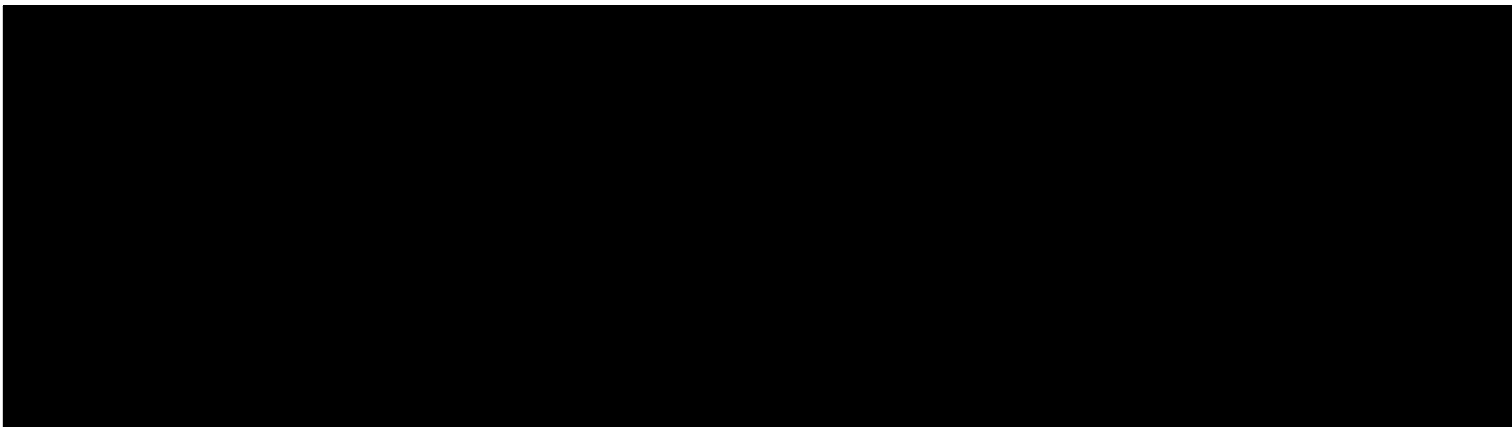
Commented [6]: why not use flogs? is there a migration plan?

Commented [7]: initially it was because we only had 20% data (that is no longer the case). We also have to flog a bunch of fields before we can switch to flogs.

Commented [8]: if you're relying on the filepusher alerts, those are somewhat unreliable.

Commented [9]: Do you have other suggestions? happy to incorporate

Commented [10]: varz checker to check if the load time is recent enough

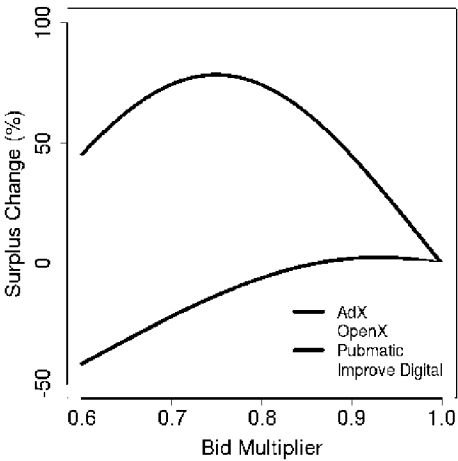


Once the above model is fit separately for each exchange, we pick the x that maximizes surplus for each advertiser.

Commented [12]: so one model per exchange?

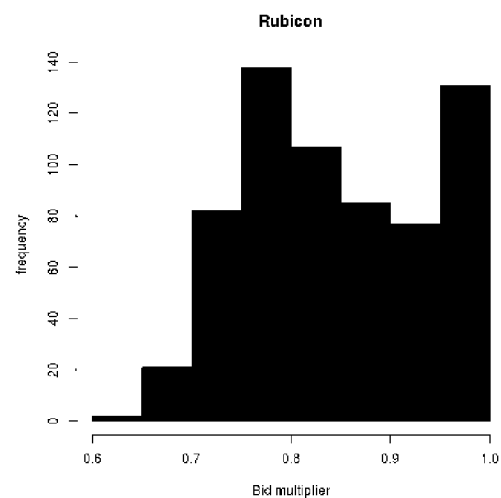
The plot below shows surplus change vs bid multiplier (x) at the exchange level (i.e., ignoring the advertiser term) for 4 exchanges.

Commented [13]: yes



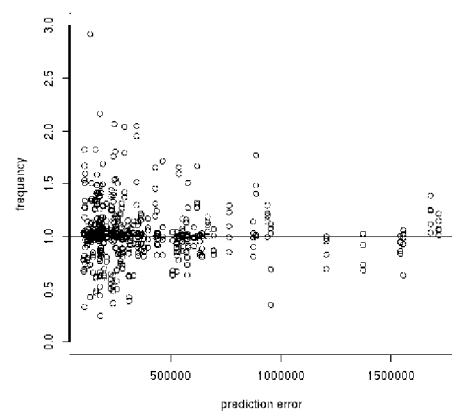
The plot below shows the histogram of bid multipliers at the advertiser level on the exchange Rubicon.

Rubicon



Cross-validation

Mean squared error = 1.1, bias = 4%



Commented [14]: why is this a good accuracy metric?

Commented [15]: it sets a benchmark. it's not bad in accuracy + it seems to work well in experiments. For future launches, when we are adding features, we can check this metric to see if quality improves

Commented [16]: fix the label

- 2. How frequently is the model updated?
Daily. Borgcron [here](#).
- 3. What is the metric for evaluating model quality?
Increase in surplus.

4. Are the models versioned?

Yes. Subscript vX to the model file names.

Serving

1. How is the model served?

Experiment file.

2. How does the model get updated in serving?

Dynamic files.

3. In case of production issues,

We can easily roll it back using an older experiment file.

4. Are there magic constants involved in serving (such as thresholds)?

If the surplus change is under 10%, we don't apply a bid multiplier

Monitoring

1. Which group is responsible for monitoring?

gtrade. gtrade-core@[google.com](mailto:gtrade-core@google.com)

2. (This is not an exhaustive list) Are there alerts for

a. Training data looks bad and/or incomplete: Yes

i. min # advertiser

ii. min # exchanges

iii. Min # background experiments

iv. Min # impressions in the previous day's data

v. Min value-added in the previous day's data

b. Trained models have poor quality: Yes. cross-validation error is checked before model is pushed.

c. Trained models are too different from the last iteration: Yes

d. Models are stale - yes, email alerts

e. Jobs failures (job down, borgcron overrun, borgcron failures) - yes, email alerts

f. Server side monitoring (see [productionisation guide for suggestion](#))

3. Is there a playbook describing how to handle the different types of alert?

No

Commented [17]: is there a pager on call rotation set up for this?

Commented [18]: and how many people are responsible for fixing these problems?

Commented [19]: it will be Tim for the most part unless he is not available, then i'll handle it. We don't have a pager for the analyst team (yet), but I heard Bahareh is working on it.

Commented [20]: Add varzs in serving

Commented [21]: how bad are stale models?

Commented [22]: I'm not terribly worried

Commented [23]: why not?

Commented [24]: Happy to add it

Production Readiness

1. How are PCRs handled?

We can easily switch to other clusters with placer deployment.